

App Analytics

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What are Analytics?

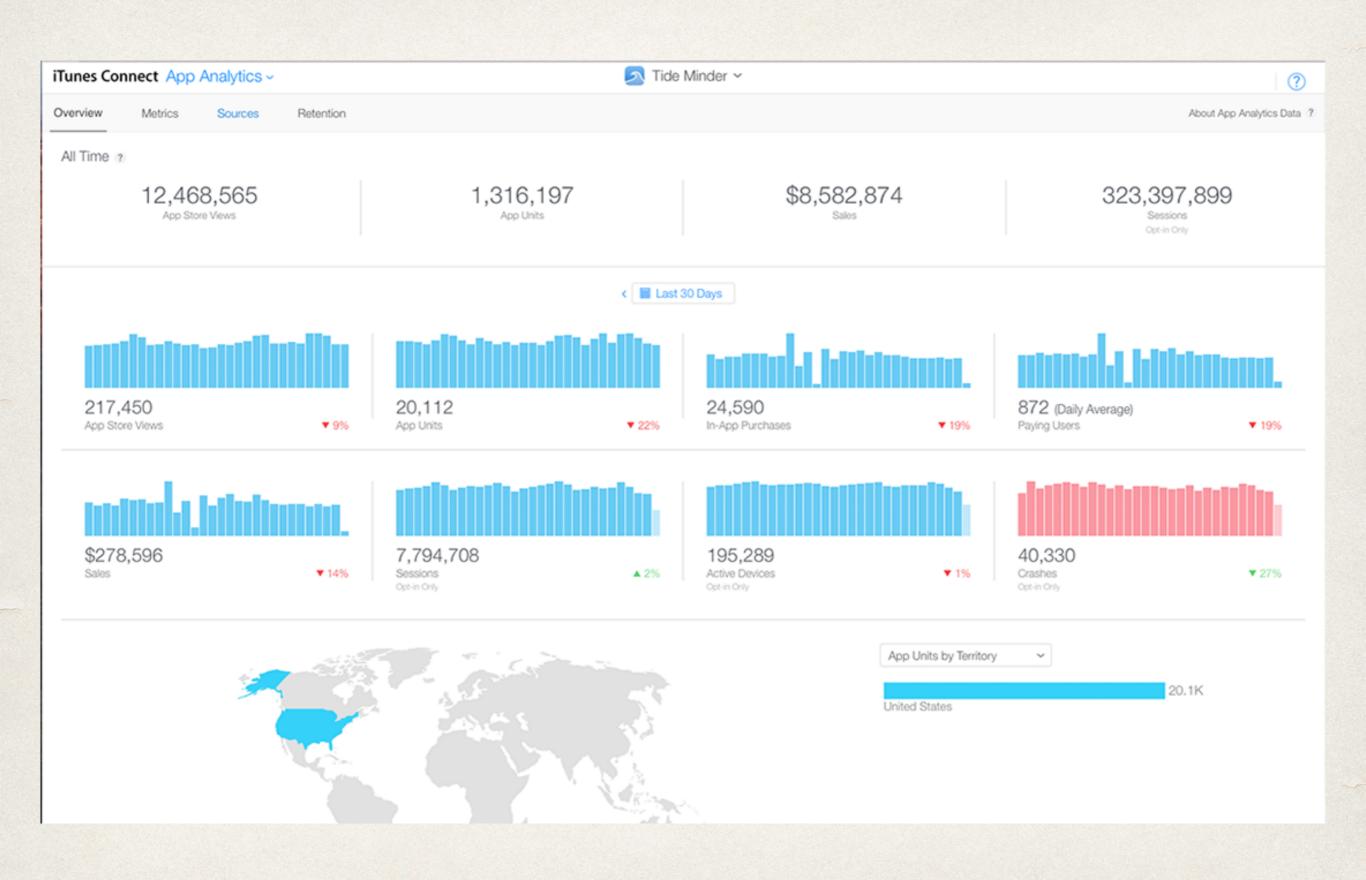
- The data-driven evaluation of user patterns
 - Data points collected based on app usage
 - Data analyzed using statistics
 - * Results visualized for human consideration
 - App patched to better serve market

iTunes Connect

- iTunes Connect provides basic analytics for all apps
 - Store views
 - Units sold
 - Active devices
 - In-app purchases
 - Crashes
- Users must opt-in for some of this shared data

Apple's App Analytics

- Data organized by:
 - App
 - Specified time period
 - Country/Region



Goals of Analytics

- Improve user experience
- Provide better marketability
- Increase profitability
- Two main metrics of success:
 - Conversion
 - Retention

Conversion

- * How many people who know about the app decide to use the app?
- Analyzes user perception of app before purchase/ download
 - Associated an app's "marketability"
- Marketing campaigns (paid advertisements or social media) can increase conversion

Conversion Metrics

- What data points might be associated with user conversion?
- Data points:
 - Product page views counts store page views
 - Impressions counts views for more than one second from any source (includes search results, featured apps etc)
 - App units sold/downloaded
 - Number of app installations

Retention

- Does the app's user base continue to use the app?
- More critical than the initial conversion!
 - "Word of mouth" is the best publicity
 - Defines reputation of both app and studio
- Good customer service, quickly addressing problems, and customer trust can increase retention

Retention Metrics

- What data points are associated with user retention?
- Data points:
 - Active devices
 - Active within last 30 days
 - Crashes
 - In-app purchases
 - Sessions

Additional Data

- General app data is only the start!
- Important to capture data related to app specific features
 - What features are most popular?
 - Where in the process are users stopping?
- Information often gathered from backend requests
 - API calls can log user data for internal analytics

Using Data

- Data doesn't say much about why only what
- Humans use data to form hypotheses about why trends are happening
- Still necessary to test trends in some statistically reasonable way

A/B Testing

- Also called split-testing
- Compare two versions of a product across multiple users
- Formalizes design decisions using actual data
- Possible to test even after deployment
- Can test for conversion or retention

How to A/B Test

- 1. Form a hypothesis about how to improve an app feature or process
 - Potential changes might be cosmetic, interaction flow, or entire features
- 2. Create variation incorporating hypothesized changes to test against the control (current version)
- 3. Deploy app versions at random across user base
 - User does not know about the testing only sees their "A" or "B" version of the app
- 4. Validate user interactions between two versions using statistical analysis



Variation B

50 % visitors

see variation B

conversion

A/B Test Example: EA



Control: 20% off future purchases

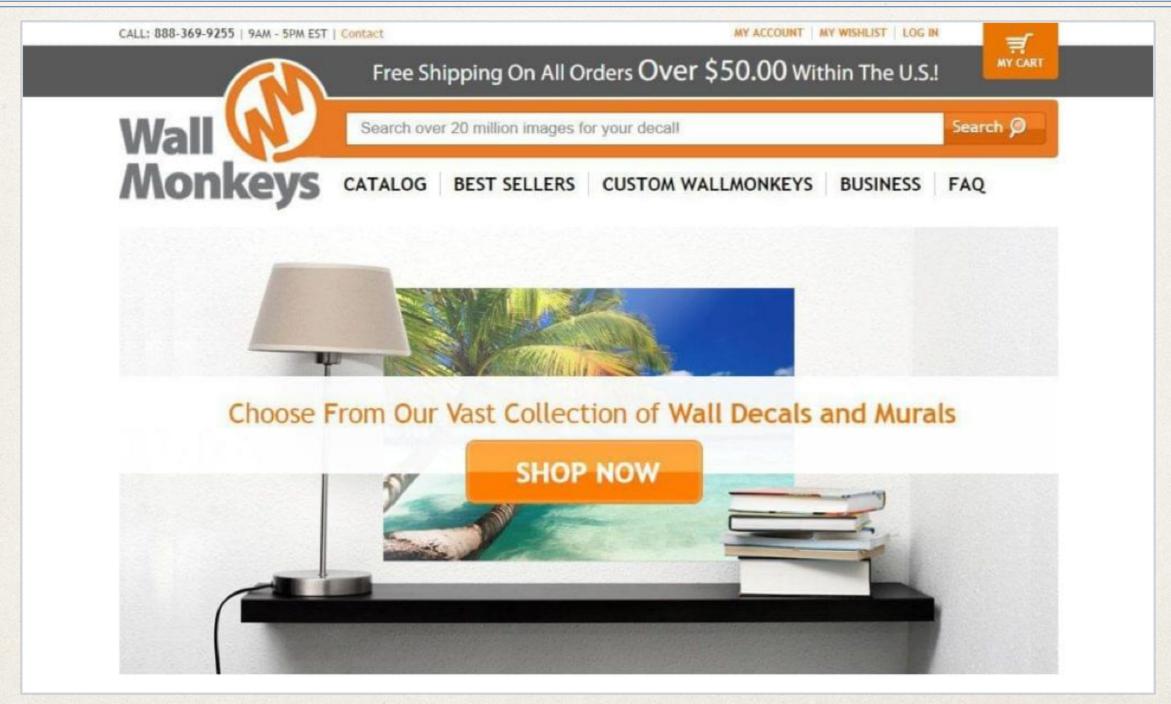
A/B Test Example: EA



Variant: No pre-order incentive

Variant or Control?

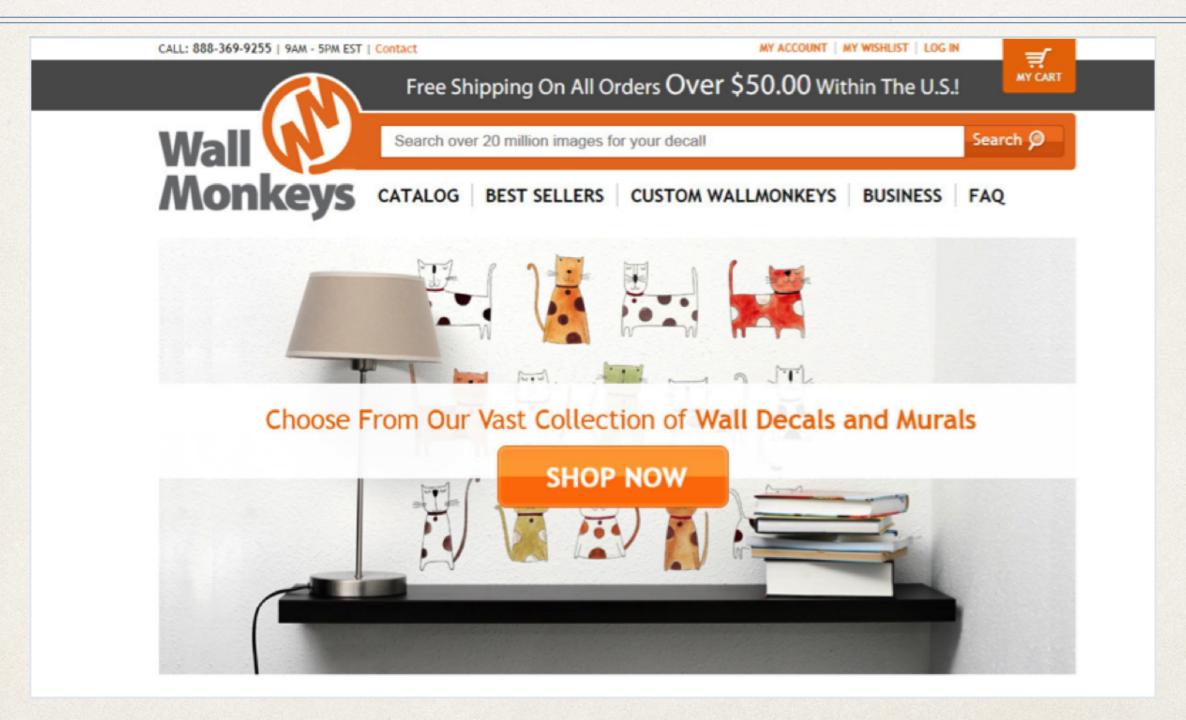
- * Variant (no pre-order incentive) performed over 40% better than control (pre-order incentive)
- Why?



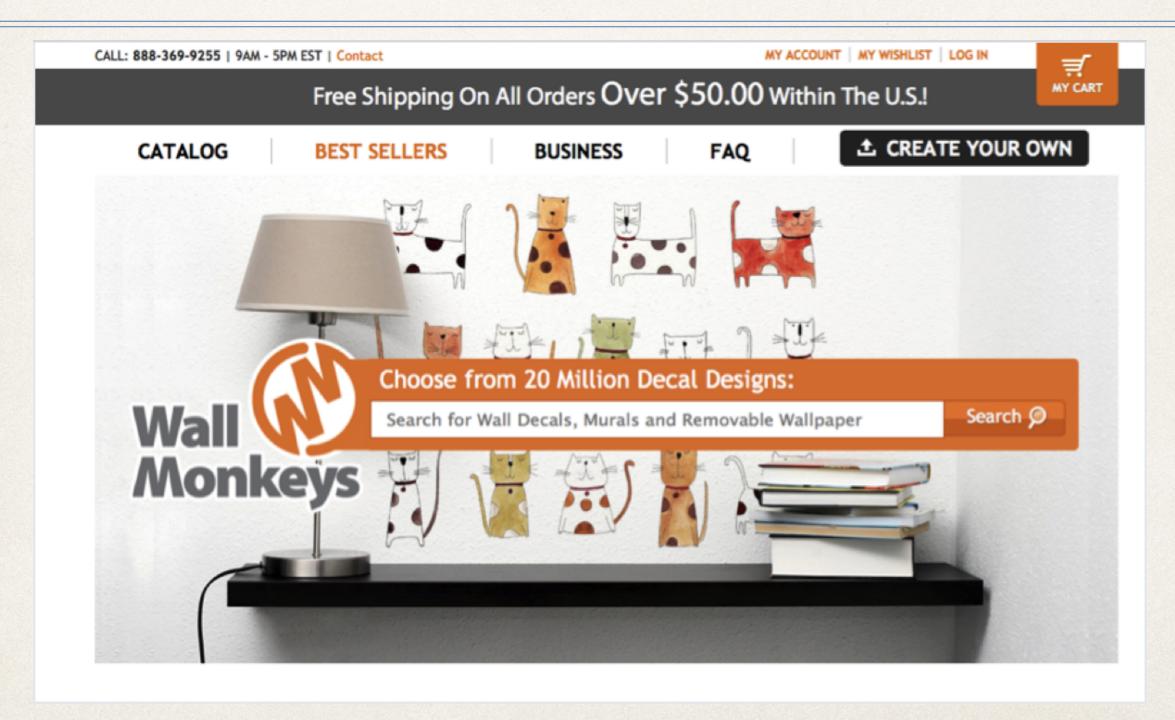
Control: original landing page



Heatmap analysis of user interactions



Variant 1: New background image



Variant 2: Updated menus and search bar

WallMonkeys Results

- Variant 1 increased conversion rate over control by 27%
- Variant 2 increased conversion rate over Variant 1 by 550%

Typical Areas of A/B Testing

- Sign-up and login flow
- Tutorials
- Visual appearance (color, size, shape, etc)

A/B Testing on Mobile

- Many services available for performing tests and app updates without resubmitting to the App Store
 - Optimizely
 - Leanplum
 - Taplytics
 - Apptimize
- Note: these tools primarily target "standard" app features such as UI/UX decisions and push notification/messaging etc

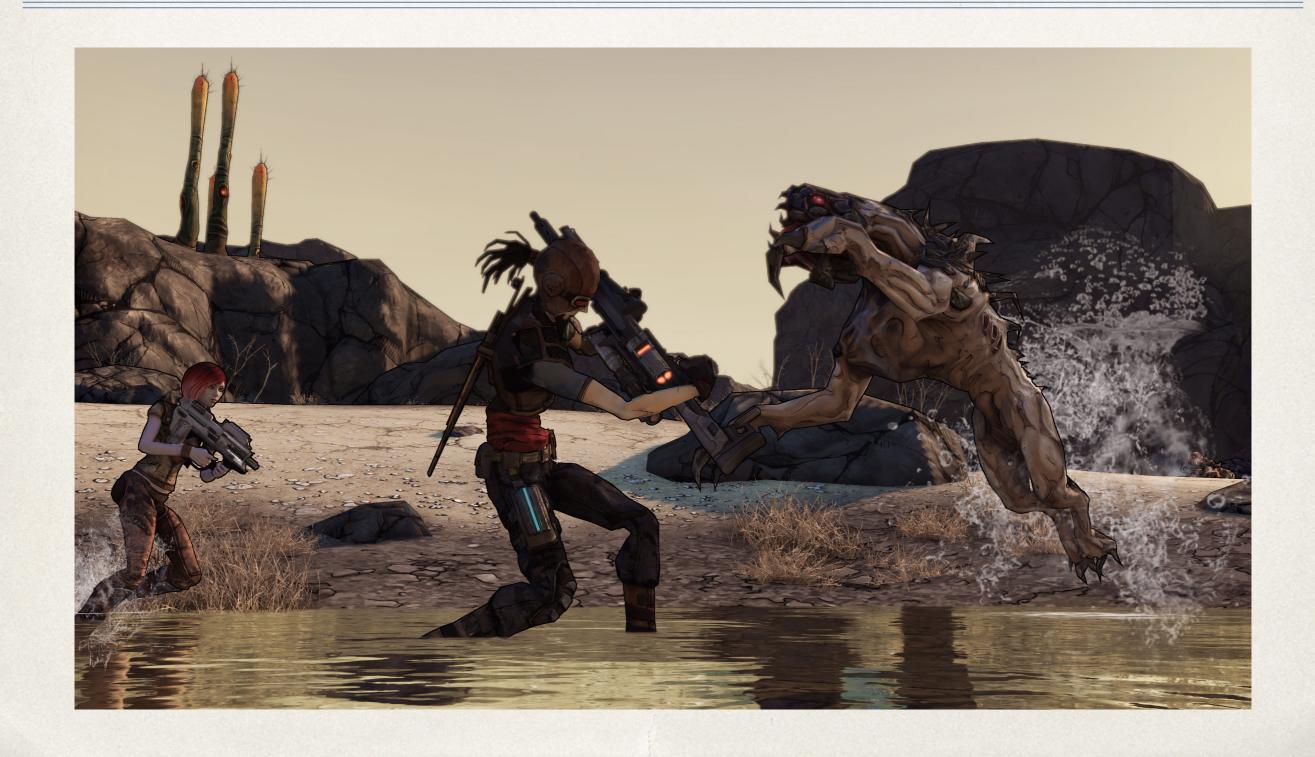
Testing Other Features

- Ongoing A/B testing works well for many web/mobile apps
- Less feasible in products like games, where "fun" and engagement are harder to define
 - Scaling of level difficulty
 - Introduction of new mechanics
 - Level/puzzle/boss design

How Can We Test These Things?

- Early play tests essential for guiding design
- Data collection of player usage can guide where fixes are necessary
 - A too-hard boss fight will lead to drop off in player base for following level
 - Preferred levels will be selected more frequently
 - Crash reports can help find bugs in levels

Example: Borderlands



Fixing Bugs and Adding Features

- Player data helps determine priority and severity of bugs and priority of features
 - Priority: How essential it is to fix or add something
 - Severity: How damaging a bug is
- Helps programmers "rank" what they should be working on

Priority vs Severity

- What is a high-priority/high-severity issue?
- What is a low-priority/low-severity issue?
- What is a high-priority / low-severity issue?
- What is a low-priority/high-severity issue?

Is Profit the Point of Data?

- Companies use data to increase profits, which is not necessarily harmful...
 - ...but could result in treating customers as a means to an end
- Data goes beyond profits and consumerism
 - Drives safety, convenience, education, happiness etc
- Consider what your app is giving users
 - If it's giving something of value, you'll get something back

Quiz Question!

* True or False (A or B): Conversion metrics include data points like crash rate and number of user sessions.

References

- * <<u>https://www.crazyegg.com/blog/ab-testing-examples/</u>>
- * <<u>https://www.raywenderlich.com/215-firebase-tutorial-ios-a-b-testing</u>>